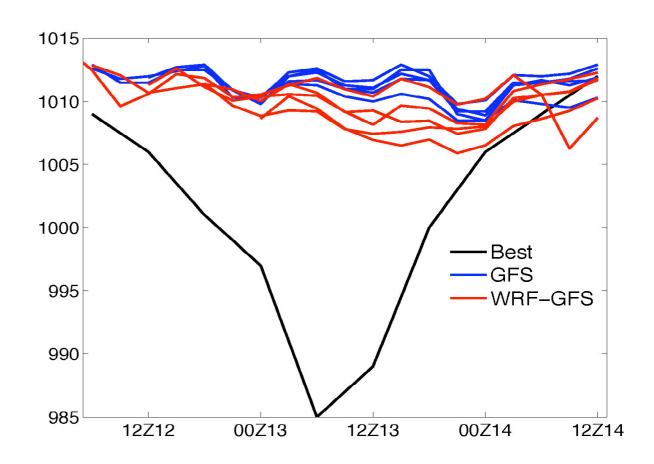
Ensemble Statistics and Error Covariance in a Rapidly Intensifying Hurricane

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Motivation

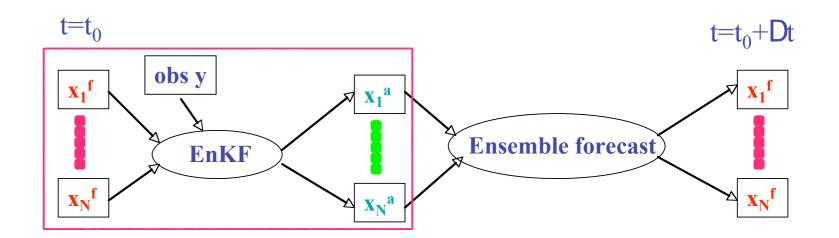
- Forecasting genesis and intensification of tropical cyclones remains challenging
 - Track forecasts vastly improved (Franklin, 2007), skill of forecasting intensity static
 - Improvements made in forecasts of other systems (Hawlblitzel et al, 2007)
- Data assimilation as a solution
 - Combines all available information optimally to best estimate state of atmosphere
 - Past studies (Zhang, 2005)
 indicate initial condition
 accuracy one key to capturing
 small, mesoscale features



GFS (**blue**) & 4.5-km WRF (**red**) forecast: No forecast initialized with GFS FNL analysis ev 6hr from 00Z 12 to 00Z 13 predicts rapid formation

Ensemble Kalman Filter

- Based on linear statistical assumptions
 - All errors assumed to be Gaussian
 - Lack of computing power prevents full solution of nonlinear systems



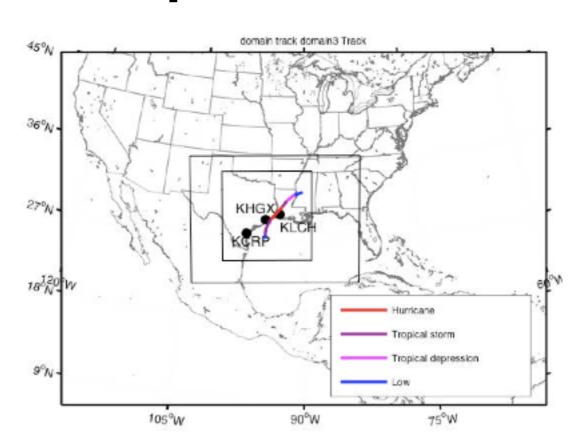
- EnKF equations
 - (1) $\mathbf{x}_a = \mathbf{x}_b + \mathbf{K}[\mathbf{y}_0 \mathbf{H}(\mathbf{x}_b)]$ Estimate of analysis
 - (2) $\mathbf{K} = \mathbf{B}\mathbf{H}^T (\mathbf{H}\mathbf{B}\mathbf{H}^T + \mathbf{R})^{-1}$ Kalman gain matrix
 - (3) $\frac{1}{N-1} \sum_{i=1}^{N} (x_i \overline{x})(y_i \overline{y})$ Covariance calculation

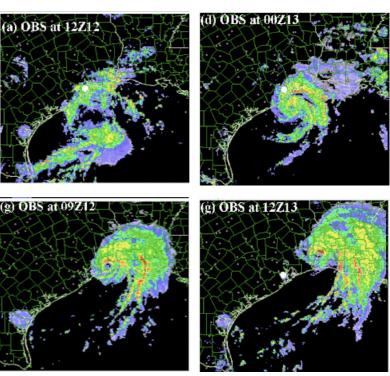
Why the EnKF?

- 3DVAR (used at NCEP)
 - Uses a mostly static, isotropic background error covariance function
- 4DVAR (used for the ECMWF)
 - Equivalent to EnKF for linear cases
- Ensemble Kalman filter
 - No adjoint model necessary (computationally efficient)
 - Dynamic background covariance
 - Background covariances estimated directly from ensemble and can capture non-linear dynamics
 - Covariance structures evolve anisotropically according to dynamics of system
 - Coherent structures most prominent in areas of moist convection, strong PV gradients (Zhang, 2005)

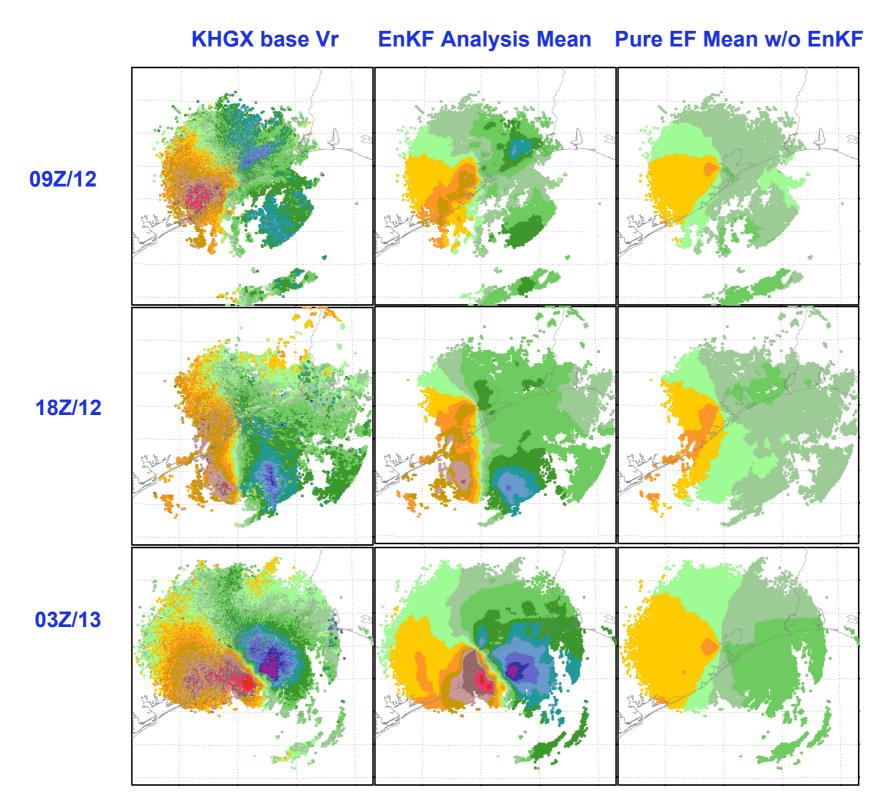
Model Setup

- WRF model used in study of Hurricane Humberto
- 3 model domains with two way nesting
- 00UTC 12 September GFS operational analysis used to create initial boundary conditions
- Ensemble size set at 30 members



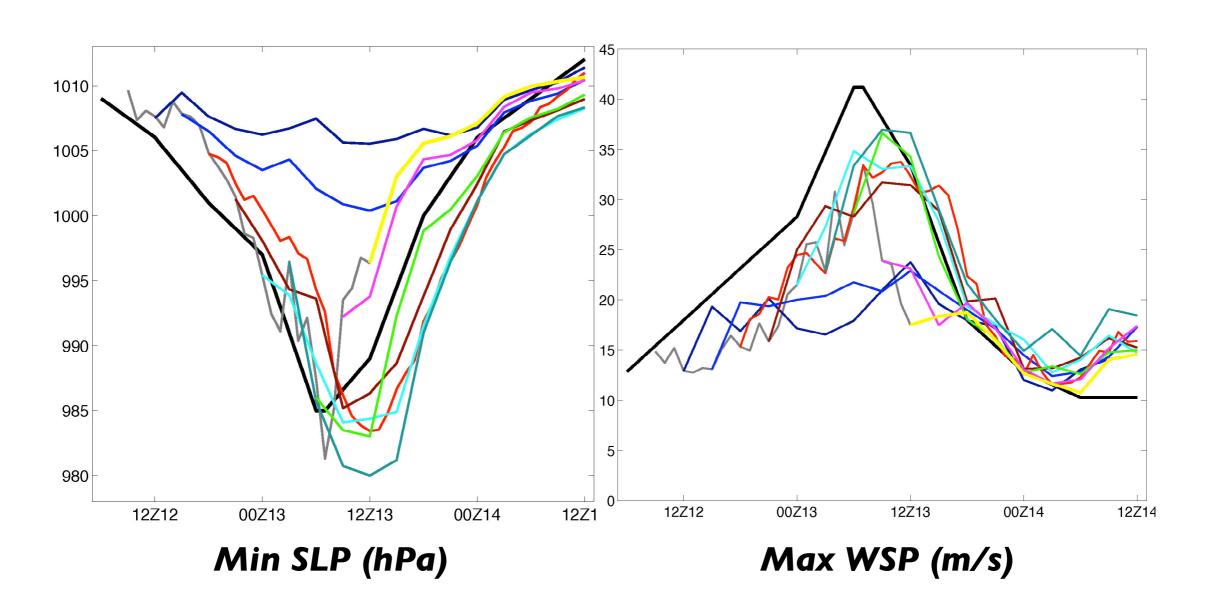


Radar Obs Assimilation



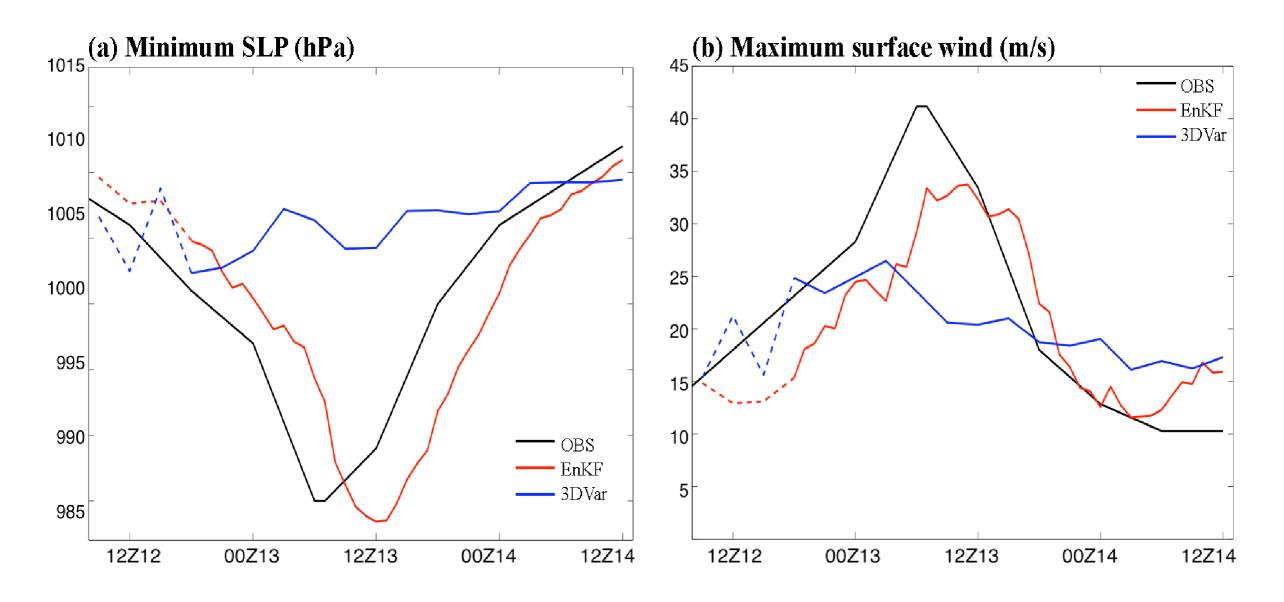
Radial velocities: Blues indicate winds toward the radar, reds indicate winds away from the radar

Radar Obs Assimilation



Forecasts initialized with EnKF analyses captured rapid cyclone deepening

WRF/3DVAR vs. WRF/EnKF

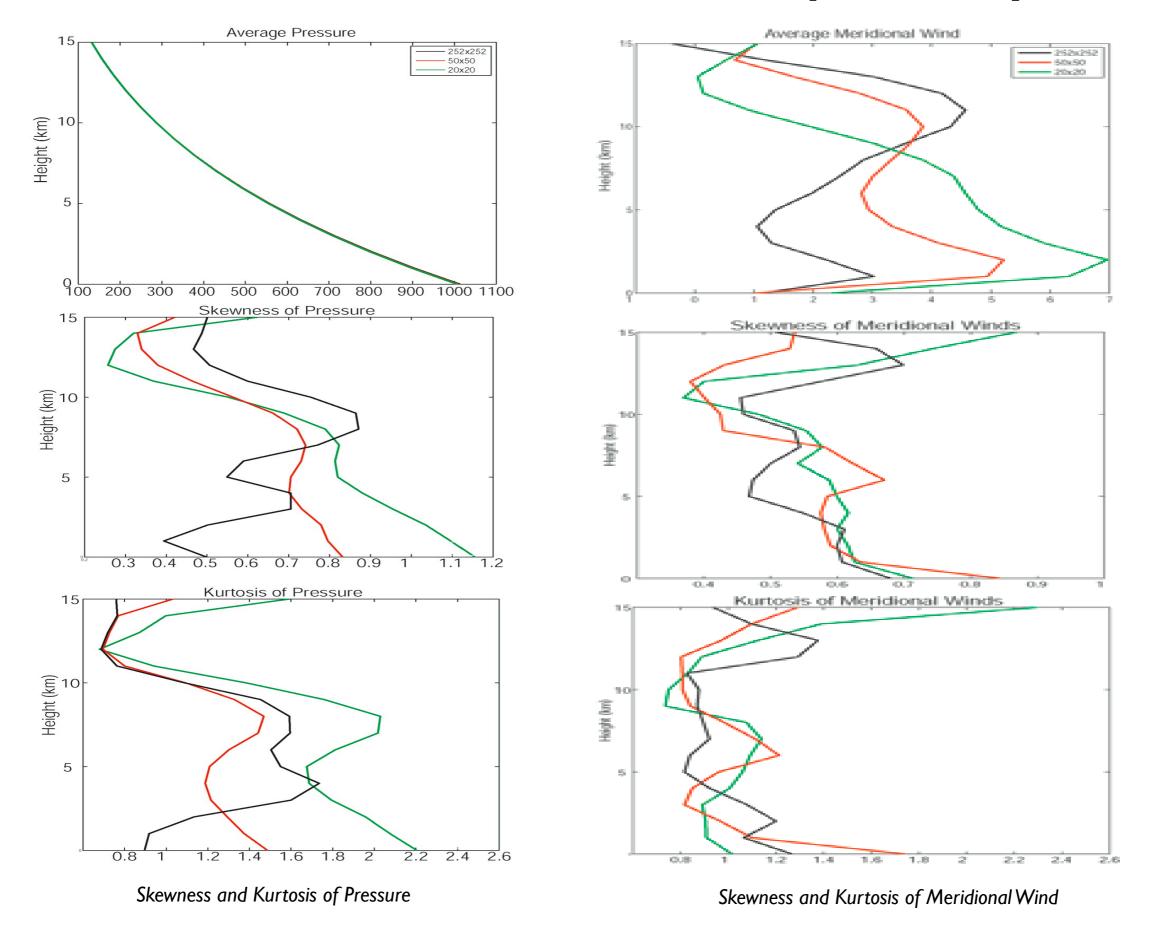


WRF/3DVAR fails to capture rapid intensification of tropical cyclone

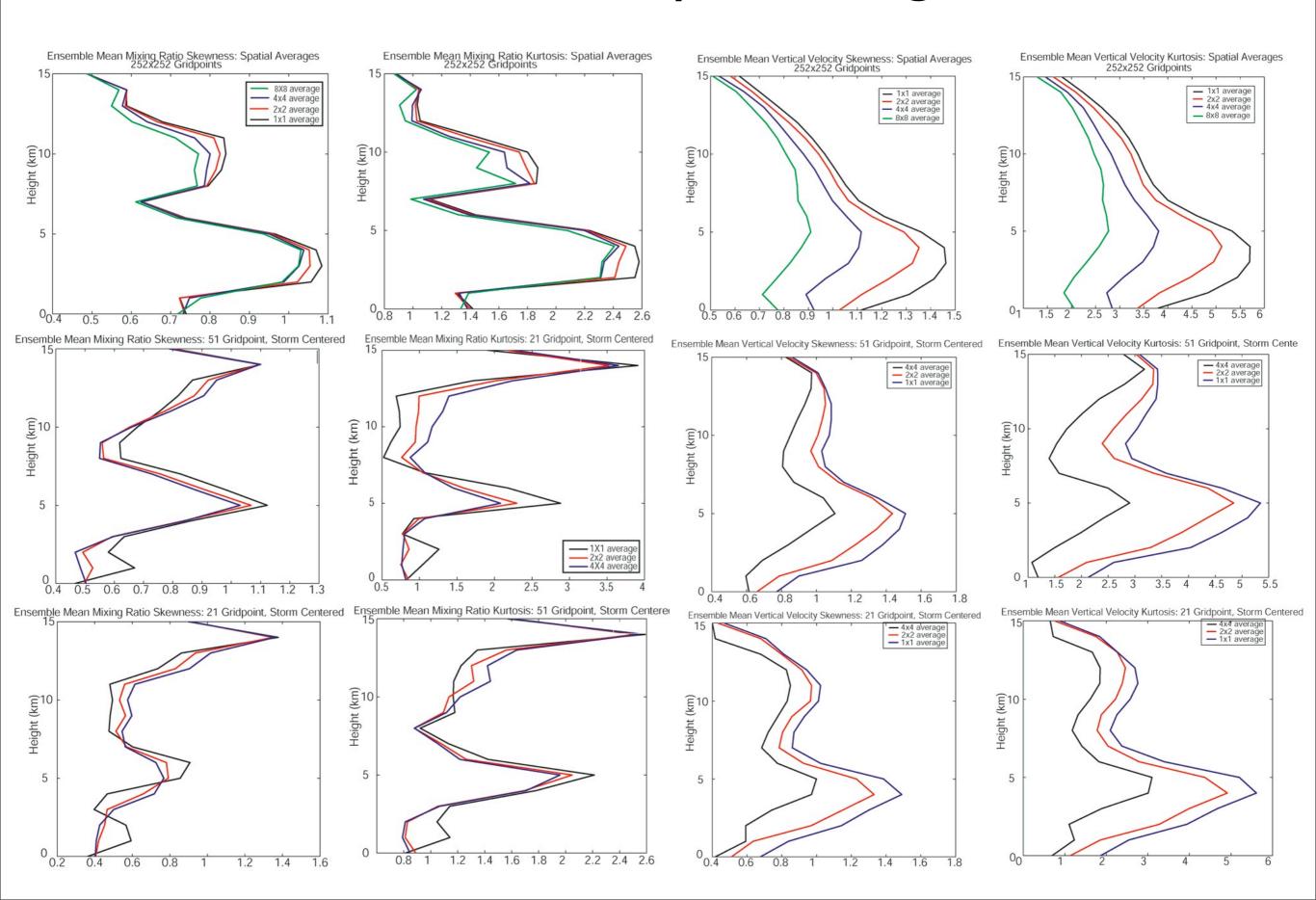
Methods

- Analysis performed at 0900 UTC on September 13 (at the time of landfall calculated by WRF ensemble)
- Statistical moments averaged across spatial domains
 - Three domains (252x252, 51x51,21x21)
 - Average taken at each vertical layer of domain (1 through 15 km)
- Will quantify non-Gaussianity of distribution and determine whether it affects performance of EnKF
- Superobs' effectiveness in reducing non-Gaussianity explored

Ensemble Statistics of a Tropical Cyclone



Effect of superobbing



Summary and Conclusions

- Current numerical models often fail to capture rapid genesis of extreme weather events
- EnKF allows flow dependent covariance calculations
- EnKF analysis and deterministic forecasts able to capture some of Humberto's rapid intensification
- Non-Gaussianity has potential to degrade EnKF updates
 - Non-Gaussianity shown to be most severe in areas of moist dynamics and intense convection
- Superobbing shows potential to perform well operationally and reduce non-Gaussianity
- Storm-centered assimilation schemes provide potential for better representation of intense mesoscale events
 - Covariances better able to be calculated near storm center